



GEOINFORMATION

in der Umweltplanung | Environmental Planning

Technische Universität Berlin



Multisource Data for ecosystem monitoring

Special Session

Mathieu Fauvel & Michael Förster

Herzlich Willkommen. Bienvenue.
Welcome.

Outline

Impact of multi-temporal research on ecosystem application

Added value of multisource classification

Requirements for efficient analysis of ecosystems by multi-temporal remote sensing



Evaluating the Temporal Stability of Synthetically Generated Time-Series



Main Question – Reasons for differences



Pre-Processing

feature selection

segmentation

filtering

Features

1st derivation
continuum removed

texture
scale-related

start of season
end of season

Target

ecosystem
inventory

ecosystem
structure

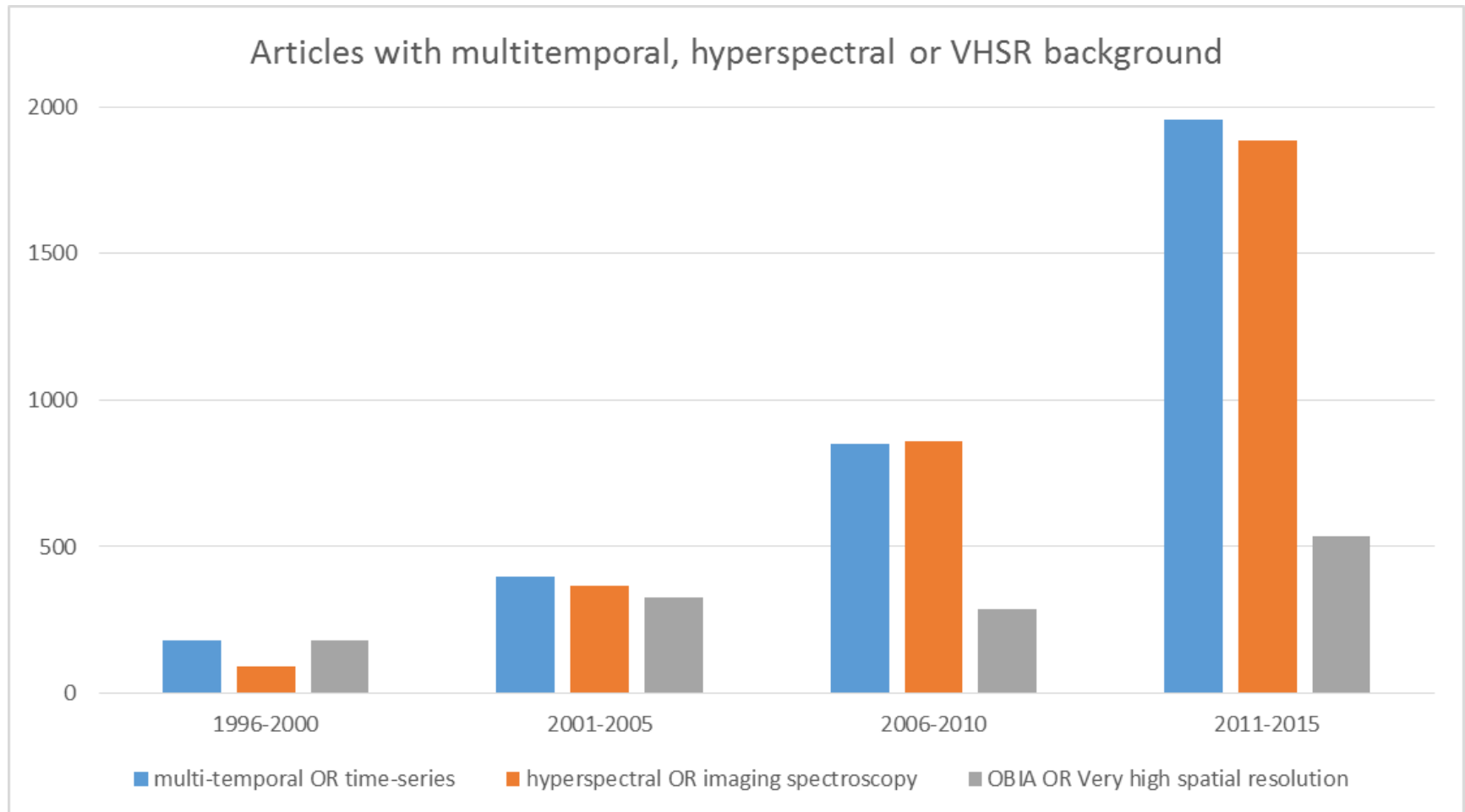
ecosystem
function

Algorithms

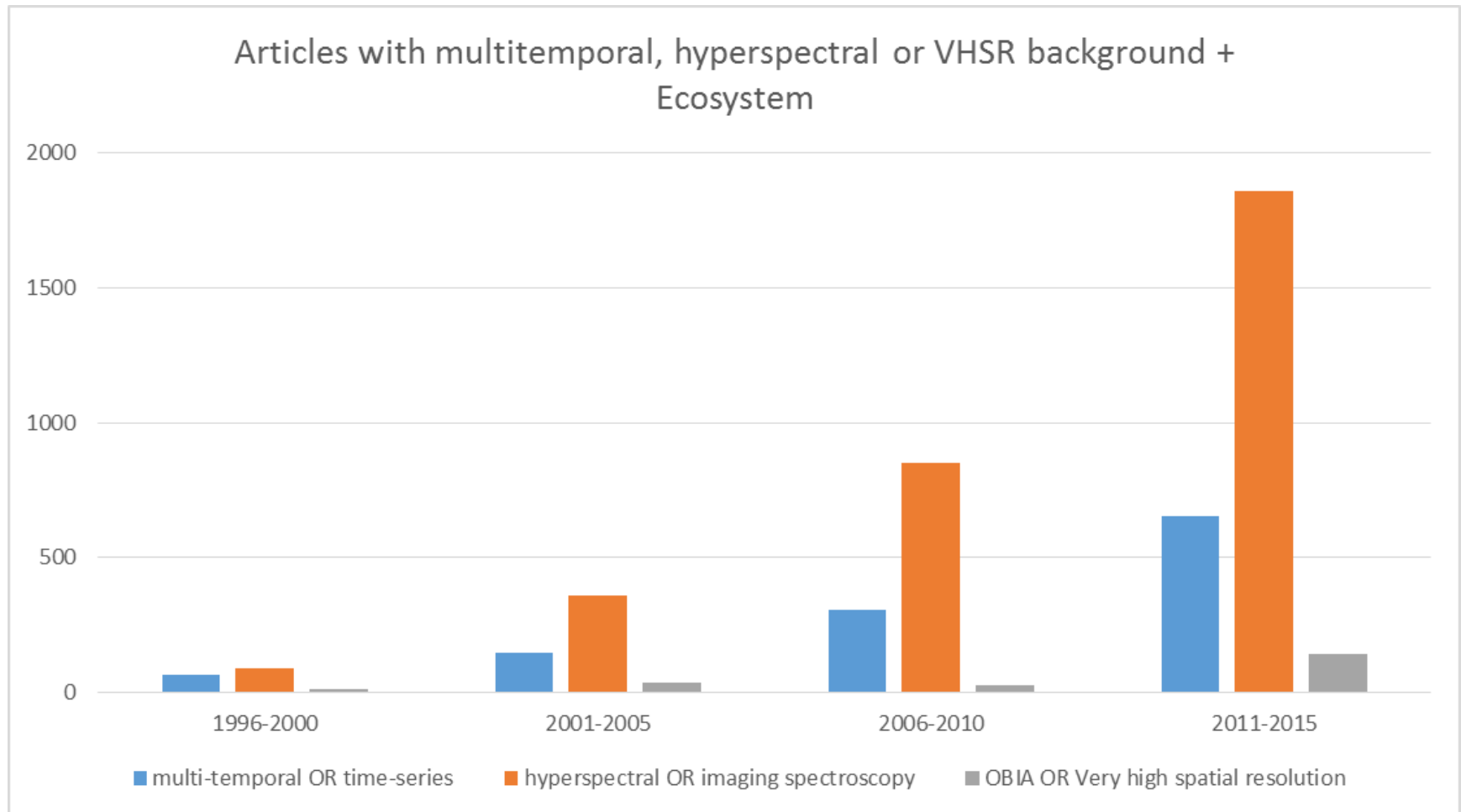
Support Vector Machines
CART-Approaches (e.g. Random Forest)



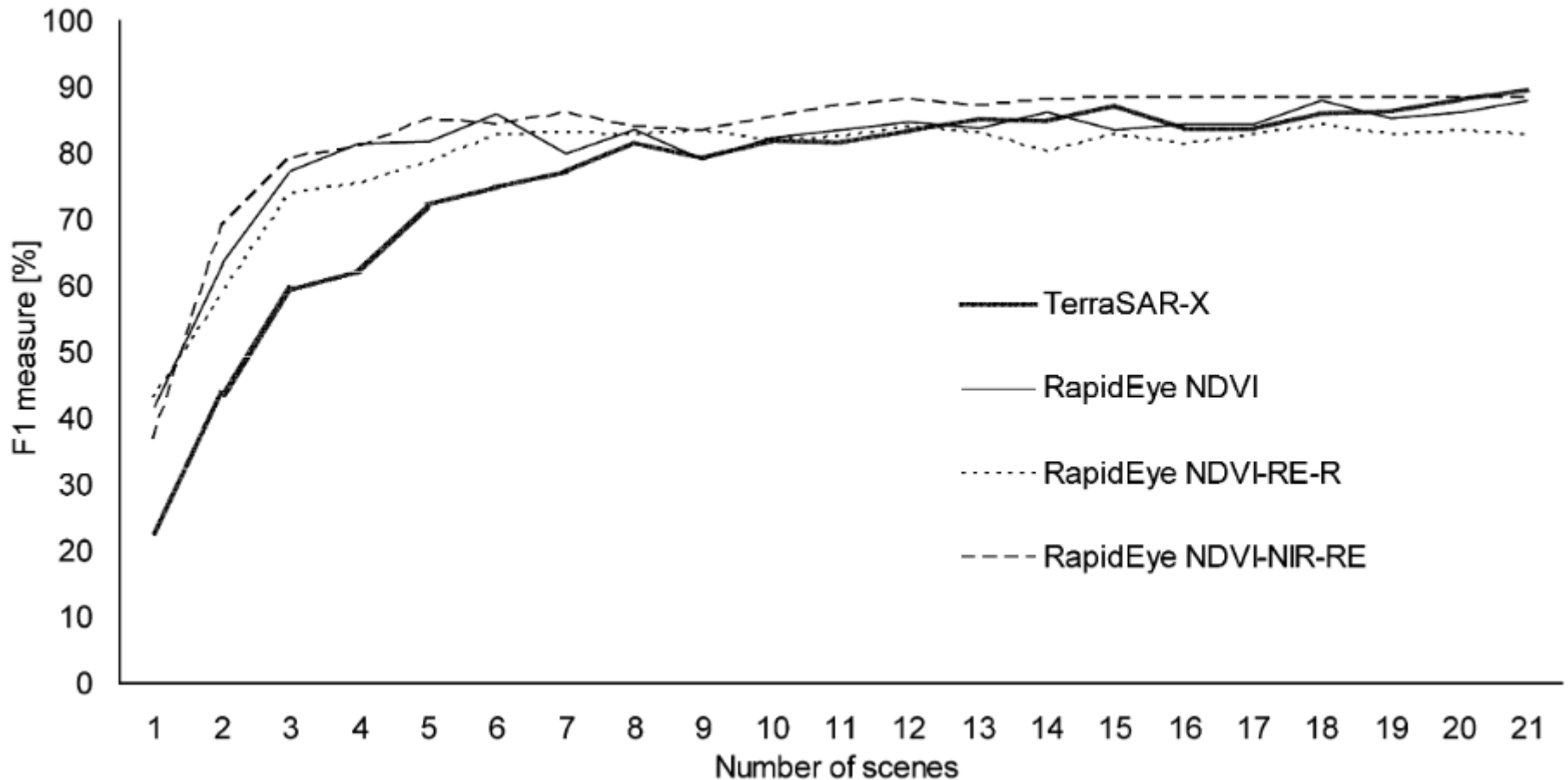
Scientific Impact of Resolutions



Scientific Impact of Resolutions



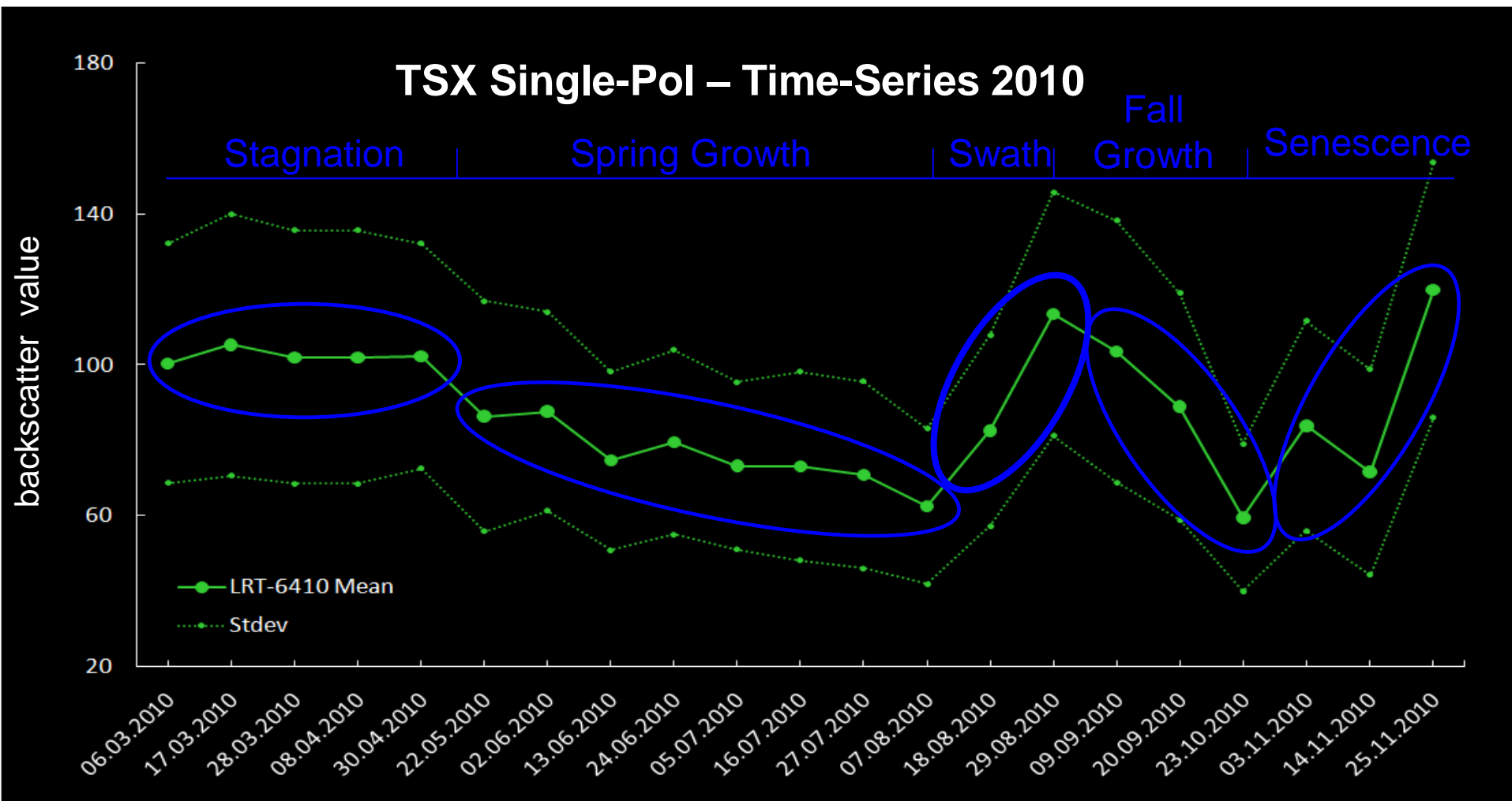
Added Value of Multitemporal & Multisource Classification



Schuster, C., Schmidt, T., Conrad, C., B., K., & Förster, M. (2015). Grassland Habitat Mapping by Intra-Annual Time Series Analysis - Comparison of RapidEye and TerraSAR-X Satellite Data. *International Journal of Applied Earth Observation and Geoinformation*, 34, 25-34



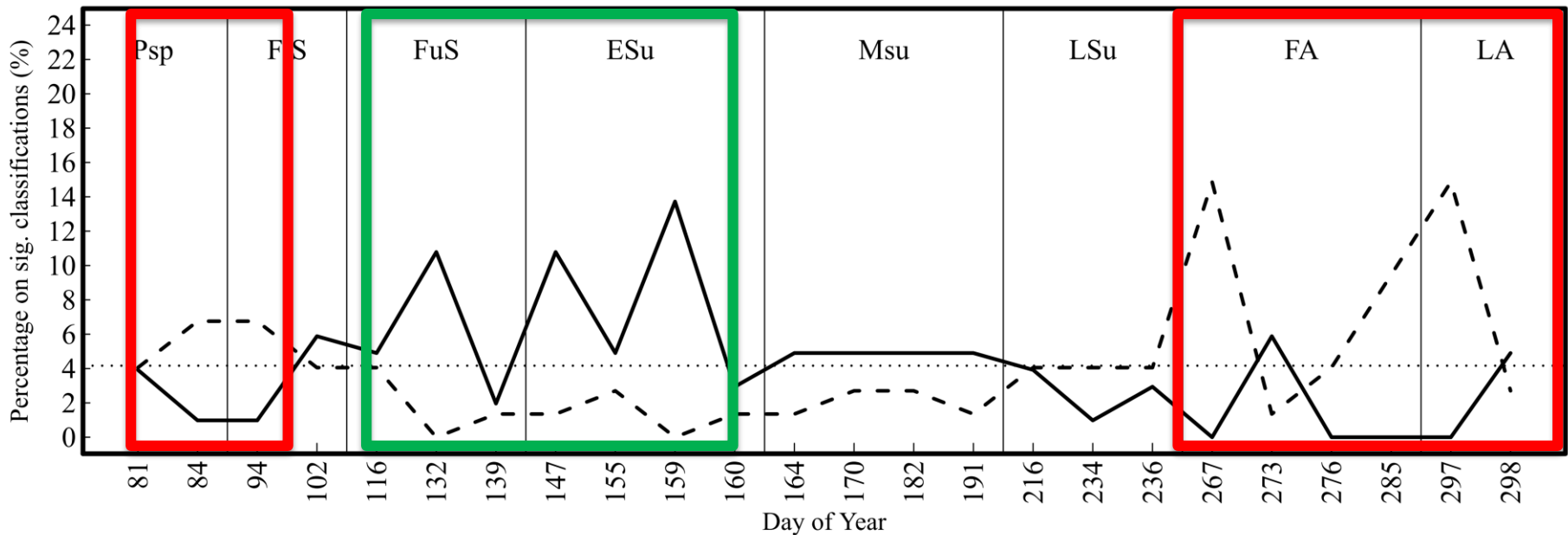
SAR Information for specific multitemporal analysis



Schuster, C., Ali, I., Lohmann, P., Frick, A., Förster, M., & Kleinschmit, B. (2011). Towards Detecting Swath Events in TerraSAR-X Time Series to Establish NATURA 2000 Grassland Habitat Swath Management as Monitoring Parameter. *Remote Sensing*, 3, 1308-1322

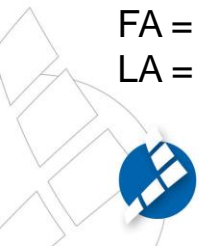


How many acquisitions and what seasonal origin are optimal?



Psp = Pre-Spring,
 FiS = First Spring,
 FuS = Full Spring,
 ESu = Early Summer,
 Msu = Midsummer,
 LSu = Late Summer,
 FA = Full Autumn,
 LA = Late Autumn

Schmidt, T., Schuster, C., Kleinschmit, B., & Förster, M. (2014). Evaluating an Intra-Annual Time Series for Grassland Classification—How Many Acquisitions and What Seasonal Origin Are Optimal? *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 7, 3428-3439



Requirements for Multitemporal Ecosystem Analysis

- large-scale coverage
- freely available

MODIS

high geometric resolution (30 m or finer)

Landsat

high repetition rate (often just a few days)

Sentinel-2

long time-series (often decades)

- availability of equidistant data (as composites)
- calibration / validation data





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Evaluating the Temporal Stability of Synthetically Generated Time-Series for Crop Types in Central Germany

**Michael Förster, Markus Möller, Feng
Gao, Tobias Schmidt, Philipp Gärtner,
Birgit Kleinschmit**

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Motivation

- temporal and spatial data fusion is still required for different ecosystem applications
- the Spatial and Temporal Adaptive Reflectance Fusion Model (STARFM) is one of the most successful ones (Gao et al. 2006)
- Chen et al. (2015) showed that a more complex landscape creates higher prediction uncertainty
- this is the case for agricultural areas

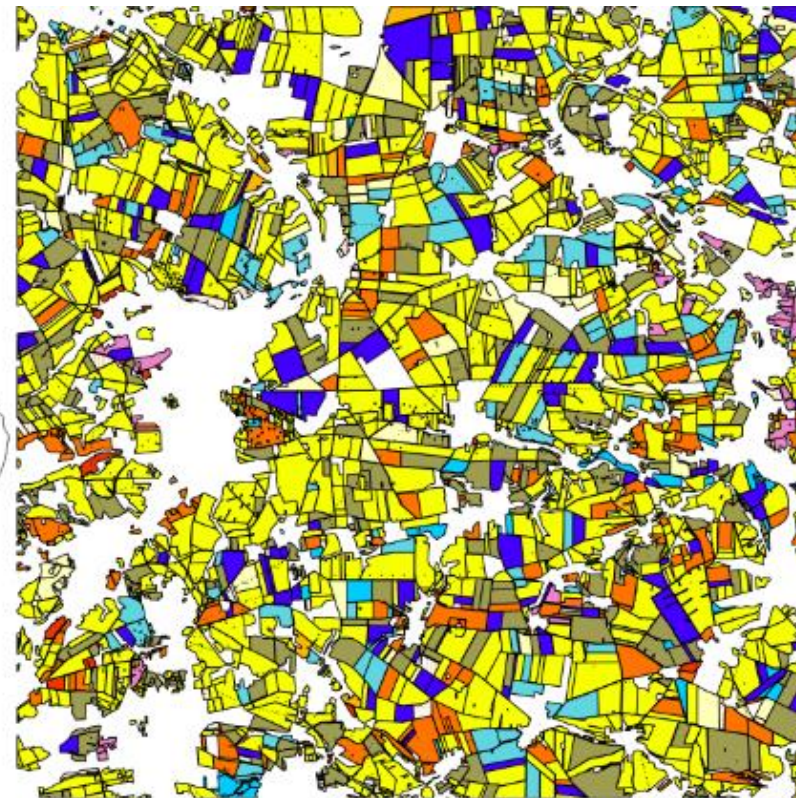
This Study is evaluating a typical STARFM product for different crops, based on a RapidEye time-series.

B. Chen, B. Huang, and B. Xu, "Comparison of Spatiotemporal Fusion Models: A Review," *Remote Sensing*, vol. 7, pp. 1798-1835, 2015.

Gao, F., Masek, J., Schwaller, M., & Hall, F. (2006). On the blending of the Landsat and MODIS surface reflectance: Predicting daily Landsat surface reflectance. *Ieee Transactions on Geoscience and Remote Sensing*, 44, 2207-2218



Study Area



CROPS 2011

- Durum (113)
- Winter Wheat (115)
- Summer Wheat (116)
- Winter Rye (121)
- Winter Barley (131)
- Summer Barley (132)
- Oat (140)
- Triticale (155)
- Corn Maize (171)
- Winter Rape (311)
- Silage Maize (411)
- Sugar Beets (620)

2.5 0 2.5 5 7.5 10 km

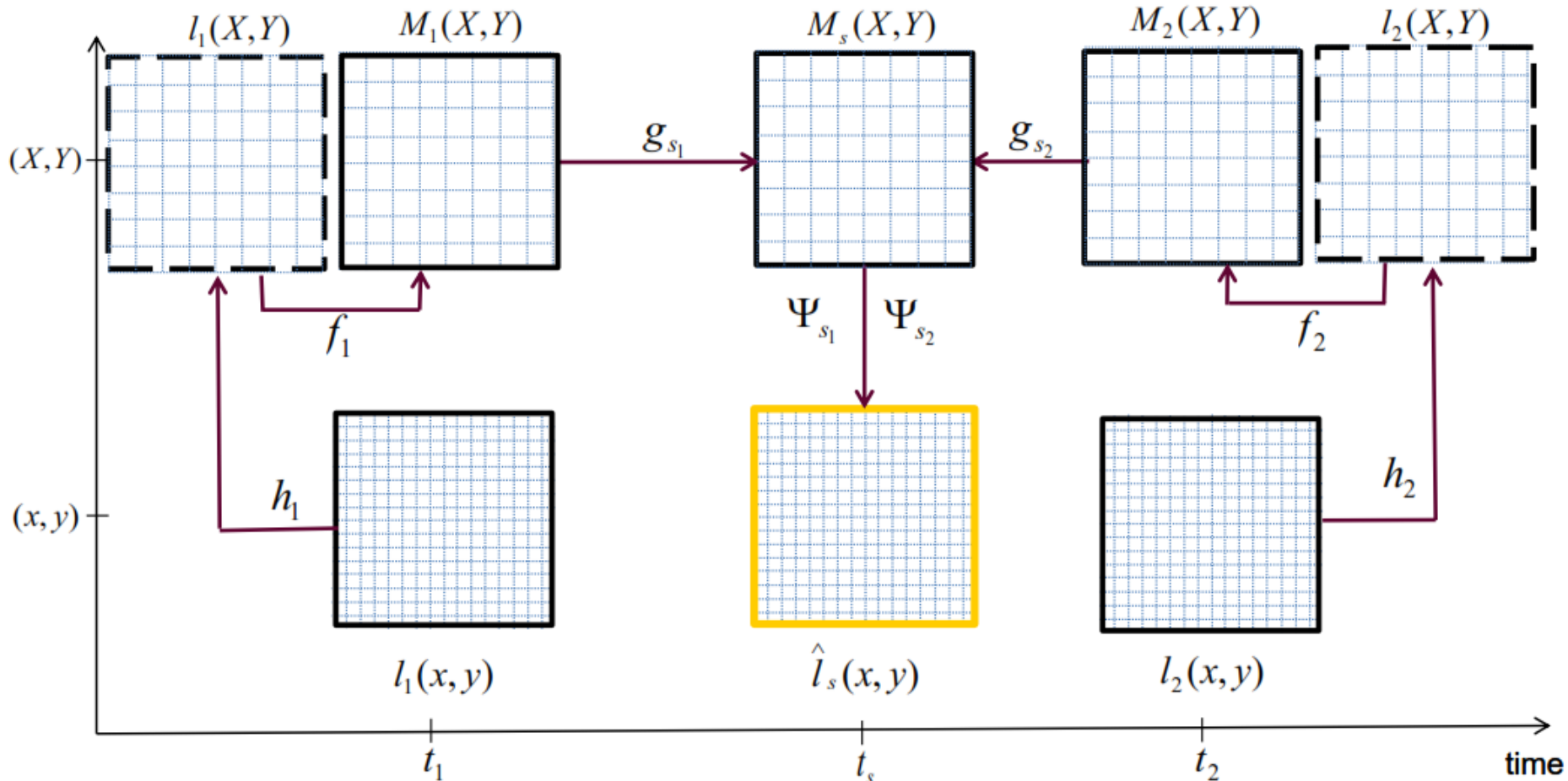
Data (for 2011)

Month	MODIS	Landsat 5/7 image (DOY) utilized for STARFM	RapidEye image (DOY)	Temporal distance (DOY)	
March	Daily product	112	66	46	
April			108	4	
			114	2	
May		128	125	3	
			128	0	
June			155	27	
			157	29	
July			208	238	30
August					
September		272	245	27	
			264	8	
October		288	317	29	
November	332		44		
	333		45		



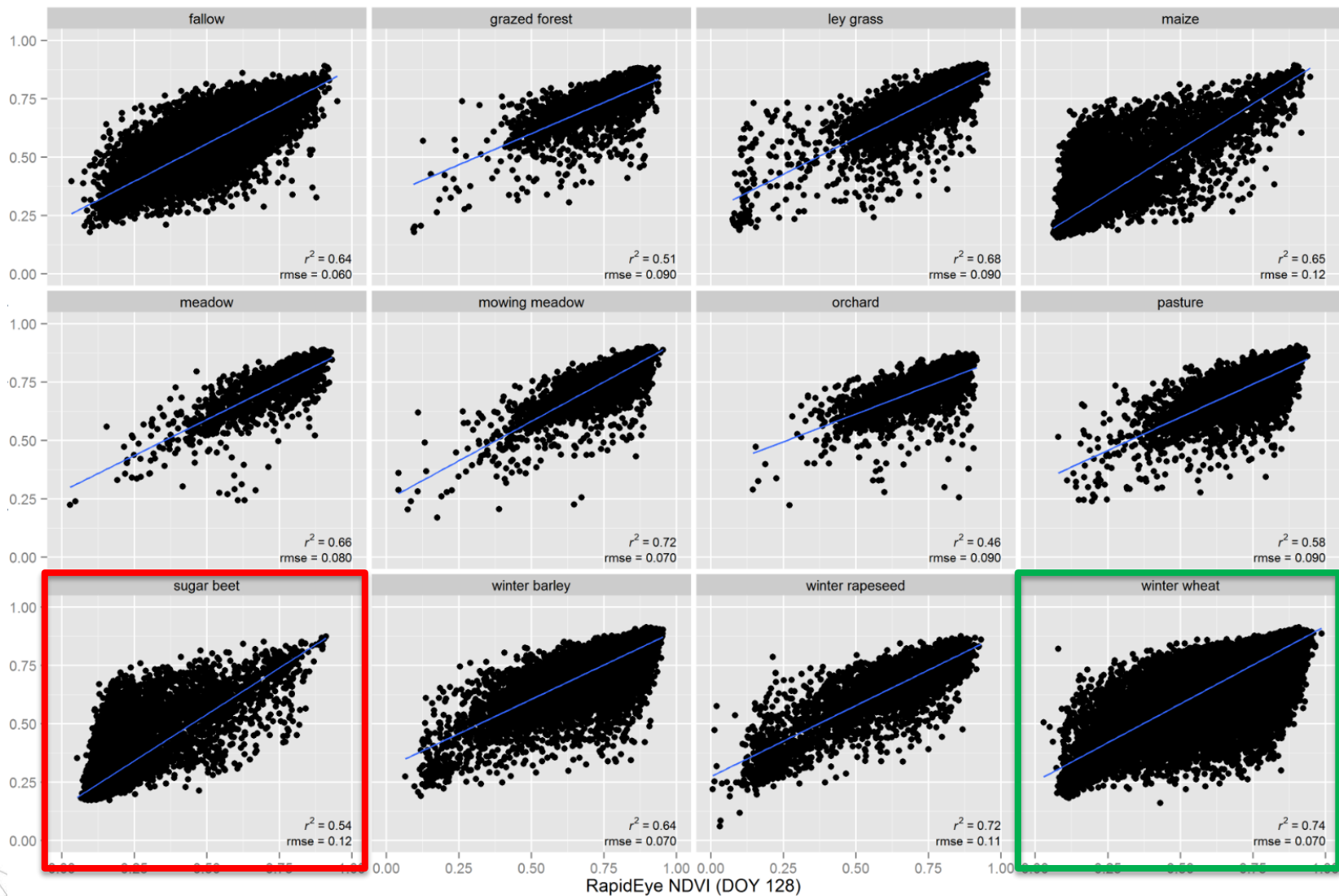
Method (STARFM)

Spatial resolution

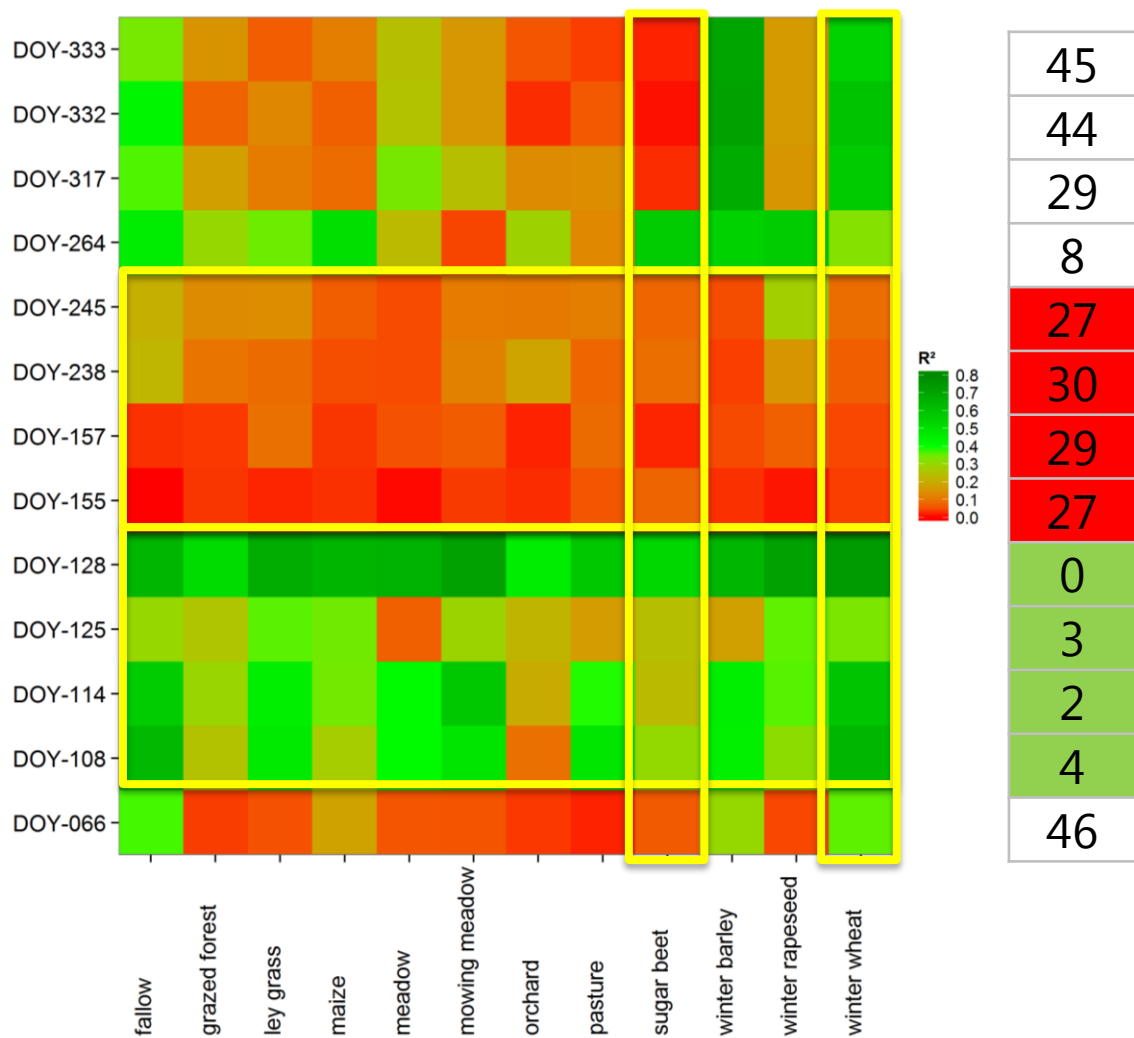


F. Gao, J. Masek, M. Schwaller, and F. Hall, "On the blending of the Landsat and MODIS surface reflectance: Predicting daily Landsat surface reflectance," *Ieee Transactions on Geoscience and Remote Sensing*, vol. 44, pp. 2207-2218, Aug 2006

Relation between synthetic Landsat and RapidEye (DOY 128)



Comparison of the synthetic scenes



45
44
29
8
27
30
29
27
0
3
2
4
46

Results depend on the date

Results depend on the crop type



Conclusion

two variables influence the quality of a synthetically derived time-series significantly:

- a) the **pairing date** of a synthetic scene should not deviate strongly from the date of interest within the time-series.

- b) **Classes with a stable phenological development** can be predicted with a higher precision compared to classes with rapid changes mostly due to multiple harvest or swath per year.





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Questions?

Michael Förster, Markus Möller, Feng Gao, Tobias Schmidt, Philipp Gärtner, Birgit Kleinschmit

Herzlichen Dank. Thank you.